

# Dataset :

## Super Store Sales analysis:

### Load Data Set :

```
df = pd.read_csv("Sample - Superstore.csv", encoding='ISO-8859-1')
df.head()
```

	Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Name	Segment	Country	City	Postal Code	Region	Product ID	Category	Sub-Category	Product Name
0	1	CA-2016-152156	11/8/2016	11/11/2016	Second Class	CG-12520	Claire Guterl	Consumer	United States	Henderson	42420	South	FUR-BO-10001798	Furniture	Bookcases	Busi Somerse Collection Bookcas
1	2	CA-2016-152156	11/8/2016	11/11/2016	Second Class	CG-12520	Claire Guterl	Consumer	United States	Henderson	42420	South	FUR-CH-10000454	Furniture	Chairs	Hon Delux Fabri Upholstere Stackin Chairs..
2	3	CA-2016-138688	6/12/2016	6/16/2016	Second Class	DV-13045	Darrin Van Huff	Corporate	United States	Los Angeles	90036	West	OFF-LA-10000240	Office Supplies	Labels	Self Adhesiv Address Labels fo Typewriter b.
3	4	US-2015-108966	10/11/2015	10/18/2015	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Fort Lauderdale	33311	South	FUR-TA-10000577	Furniture	Tables	Bretfor CR450 Series Slin Rectangula

### Perform Task :

#### 1. Shape

```
df.shape # how much row and column
```

```
(9994, 21)
```

#### 2. Columns

```
df.columns # Columns name
```

```
Index(['Row ID', 'Order ID', 'Order Date', 'Ship Date', 'Ship Mode',  
      'Customer ID', 'Customer Name', 'Segment', 'Country', 'City', 'State',  
      'Postal Code', 'Region', 'Product ID', 'Category', 'Sub-Category',  
      'Product Name', 'Sales', 'Quantity', 'Discount', 'Profit'],  
      dtype='object')
```

### 3. Information of Data

```
df.info() # table information
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
Data columns (total 21 columns):
#   Column              Non-Null Count  Dtype
---  -
0   Row ID              9994 non-null   int64
1   Order ID            9994 non-null   object
2   Order Date          9994 non-null   object
3   Ship Date           9994 non-null   object
4   Ship Mode           9994 non-null   object
5   Customer ID         9994 non-null   object
6   Customer Name       9994 non-null   object
7   Segment            9994 non-null   object
8   Country             9994 non-null   object
9   City               9994 non-null   object
10  State              9994 non-null   object
11  Postal Code        9994 non-null   int64
12  Region            9994 non-null   object
13  Product ID         9994 non-null   object
14  Category           9994 non-null   object
15  Sub-Category       9994 non-null   object
16  Product Name       9994 non-null   object
17  Sales              9994 non-null   float64
18  Quantity           9994 non-null   int64
19  Discount           9994 non-null   float64
20  Profit             9994 non-null   float64
dtypes: float64(3), int64(3), object(15)
memory usage: 1.6+ MB
```

### 4. Duplicate value

```
df.duplicated().sum() # Find duplicate in given data
np.int64(0)
```

### 5. Business Performance

```
# Total sales , profit, orders , unique customers
total_sales = df['Sales'].sum()
total_profits = df['Profit'].sum()
unique_customers = df['Customer Name'].nunique()
total_orders = df['Order ID'].nunique()
df.head(793)

total_sales , total_profits , unique_customers, total_orders

(np.float64(2297200.8603000003), np.float64(286397.0217), 793, 5009)
```

### 6. Maximum Sales

```
# find maximum sales
df.groupby('Category')['Sales'].max()

Category
Furniture      4416.174
Office Supplies 9892.740
Technology     22638.480
Name: Sales, dtype: float64
```

### 7. Minimum Sales

```
# find minimum sales
df.groupby('Category')['Sales'].min()

Category
Furniture      1.892
Office Supplies 0.444
Technology     0.990
Name: Sales, dtype: float64
```

### 8. Total sales by Year and Month

```
# aggregates total sales by Year and Month
df['Year'] = df['Order Date'].dt.year
df['Month'] = df['Order Date'].dt.month_name()
df.groupby('Year')['Sales'].sum()
df.groupby('Month')['Sales'].sum().sort_values(ascending = False)

Month
November      352461.0710
December      325293.5035
September     307649.9457
March         205005.4888
October       200322.9847
August        159044.0630
May           155028.8117
June          152718.6793
July          147238.0970
April         137762.1286
January       94924.8356
February      59751.2514
Name: Sales, dtype: float64
```

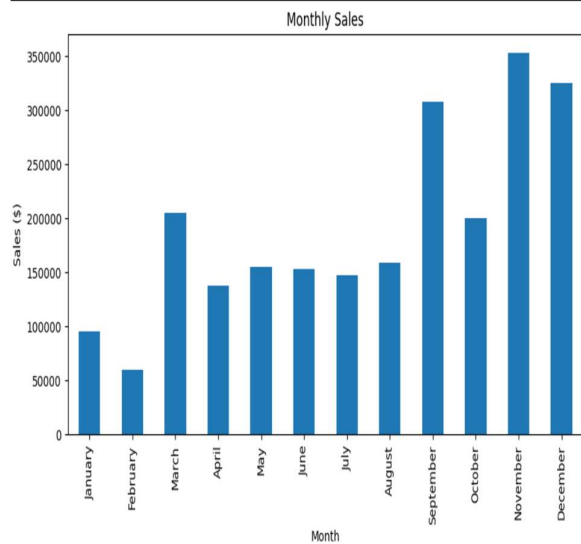
### 9. Finds the mean profit

```
# finds the mean profit
df.groupby('Region')['Profit'].mean()

Region
Central      17.092709
East         32.135808
South        28.857673
West         33.849032
Name: Profit, dtype: float64
```

## 10. Monthly sales using plot

```
# monthly sales using plot
monthly_sales = df.groupby('Month')['Sales'].sum()
monthly_sales = monthly_sales.reindex([
    'January', 'February', 'March', 'April', 'May', 'June',
    'July', 'August', 'September', 'October', 'November', 'December'
])
monthly_sales.plot(kind='bar', figsize=(10,5), title='Monthly Sales')
plt.ylabel('Sales ($)')
plt.show()
```



## 11. Generate highest profit

```
# generate the highest overall profit
df.groupby('Sub-Category')['Profit'].sum().sort_values(ascending = False)
```

Sub-Category	Profit
Copiers	55617.8249
Phones	44515.7386
Accessories	41936.6357
Paper	34853.5693
Binders	38221.7633
Chairs	26590.1663
Storage	21278.8264
Appliances	18138.8854
Furnishings	13859.1436
Envelopes	6964.1767
Art	6527.7878
Labels	5546.2548
Machines	3384.7548
Fasteners	949.5182
Supplies	-1189.8895
Bookcases	-3472.5568
Tables	-17725.4811

Name: Profit, dtype: float64

## 11. Generate highest profit

```
# generate the highest overall profit
df.groupby('Sub-Category')['Profit'].sum().sort_values(ascending = False)

Sub-Category
Copiers      55617.8249
Phones       44515.7306
Accessories  41936.6357
Paper        34053.5693
Binders      30221.7633
Chairs       26590.1663
Storage      21278.8264
Appliances   18138.0054
Furnishings  13059.1436
Envelopes    6964.1767
Art          6527.7870
Labels       5546.2540
Machines     3384.7569
Fasteners    949.5182
Supplies    -1189.0995
Bookcases   -3472.5560
Tables      -17725.4811
Name: Profit, dtype: float64
```

## 12. use for date like [ year – month- day]

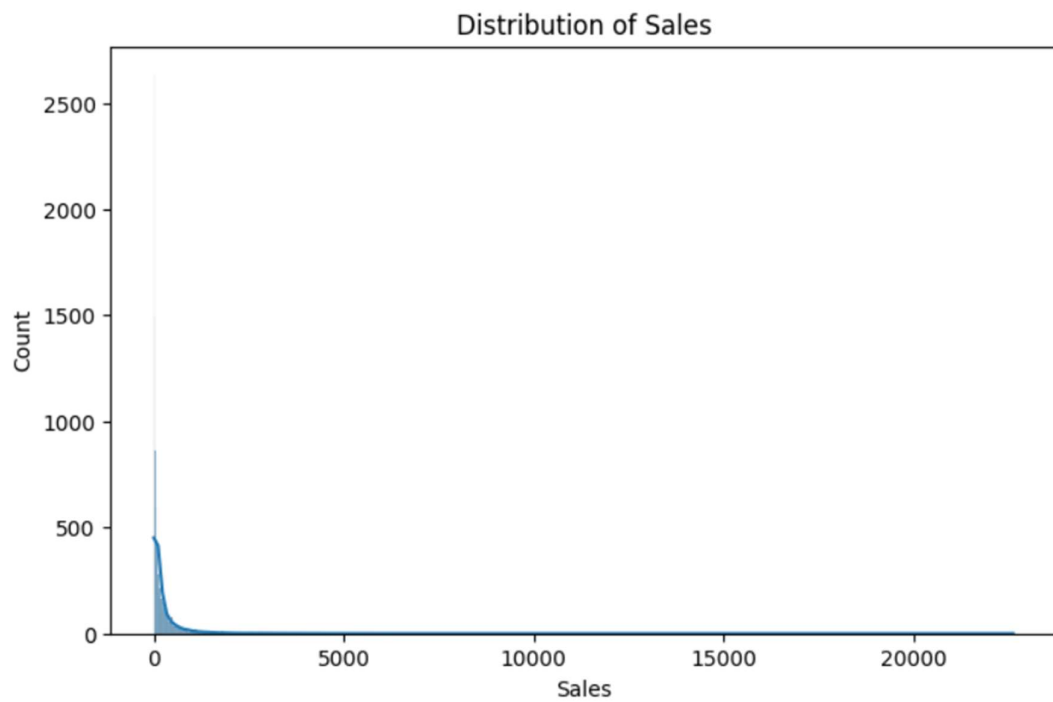
```
# Use for date like [ year-month-day]
df['Order Date'] = pd.to_datetime(df['Order Date'])
df['Ship Date'] = pd.to_datetime(df['Ship Date'])
df.head(2) # Use for date like [ year-month-day]
```

	Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Name	Segment	Country	City	...	Postal Code	Region	Product ID	Category
0	1	CA-2016-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderson	...	42420	South	FUR-BO-10001798	Furniture
1	2	CA-2016-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderson	...	42420	South	FUR-CH-10000454	Furniture

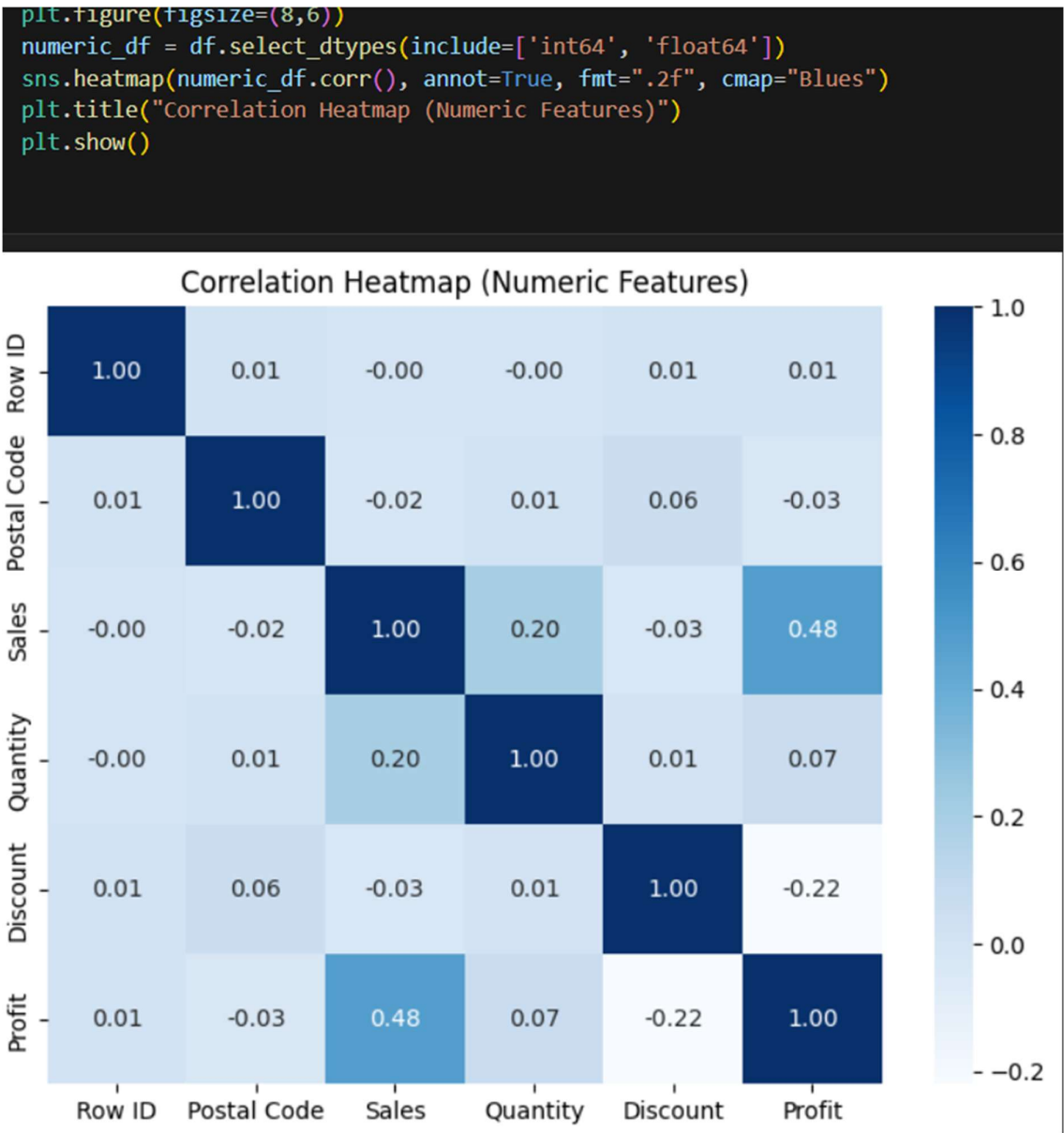
2 rows x 21 columns

### 13. Distribution of Sales

```
plt.figure(figsize=(8,5))  
sns.histplot(df['Sales'], kde=True)  
plt.title("Distribution of Sales")  
plt.show()
```



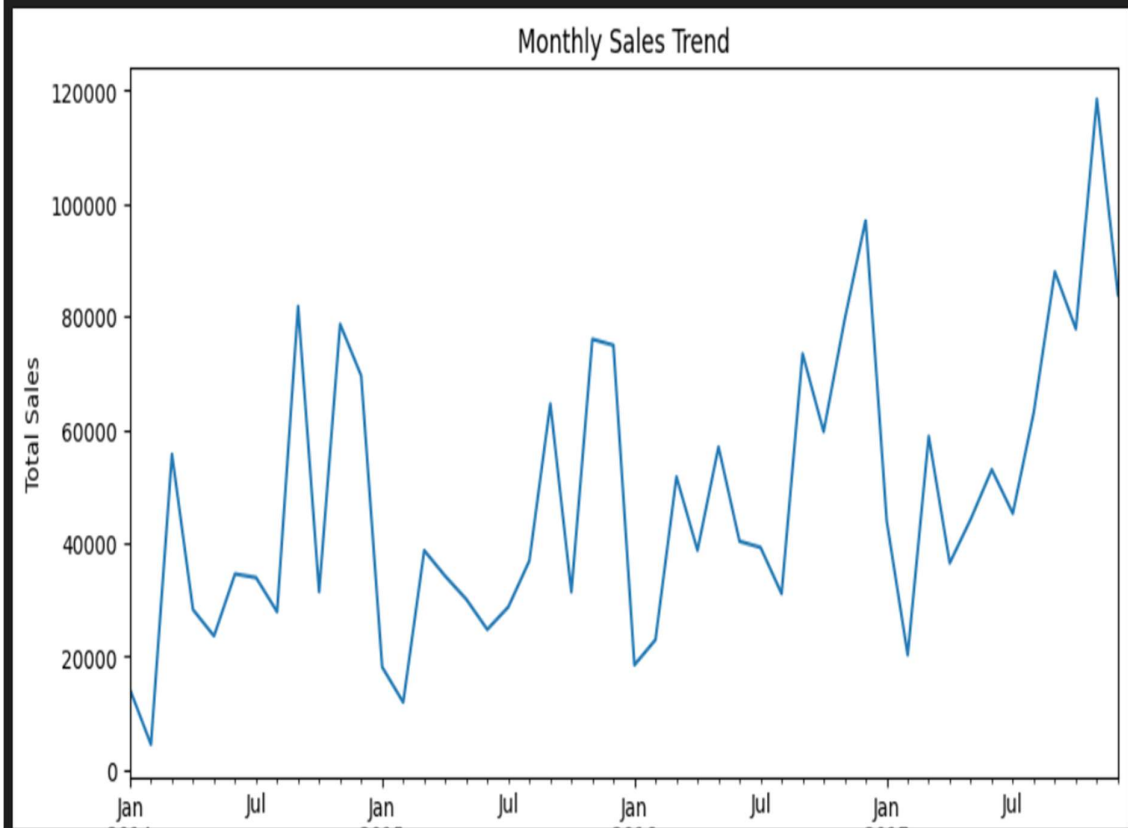
14. Correlation Heatmap(Numeric Features)



## 15. Monthly Sales Trend

```
df_time = df.groupby(df["Order Date"].dt.to_period("M"))["Sales"].sum()

plt.figure(figsize=(10,5))
df_time.plot()
plt.title("Monthly Sales Trend")
plt.xlabel("Year-Month")
plt.ylabel("Total Sales")
plt.show()
```



## 16. Compare Sales Vs Profit

```
plt.figure(figsize=(7,5))
sns.scatterplot(x="Sales", y="Profit", data=df)
plt.title("Sales vs Profit")
plt.show()
```





## 17. Total Profit by Segment

```
plt.figure(figsize=(7,5))
sns.barplot(x="Segment", y="Profit", data=df, estimator=sum, palette="viridis")
plt.title("Total Profit by Segment")
plt.show()
```

C:\Users\tarpi\AppData\Local\Temp\ipykernel\_1964\105888946.py:2: FutureWarning:

Passing 'palette' without assigning 'hue' is deprecated and will be removed in v0.14.0. Assign the 'x' variable to 'hue' and set 'legend=False' for the same effect.

```
sns.barplot(x="Segment", y="Profit", data=df, estimator=sum, palette="viridis")
```

